

# **Bayes Days 2000 at LANL**

## **Three-Day Minicourse on Bayesian Analysis in Physics**

Lectures presented by

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Max Planck Institute for Plasma Physics

Sponsored by

Enhanced Surveillance Program, Los Alamos National Laboratory

For more information, look on the web:  
<http://public.lanl.gov/kmh/course/BD2000.html>

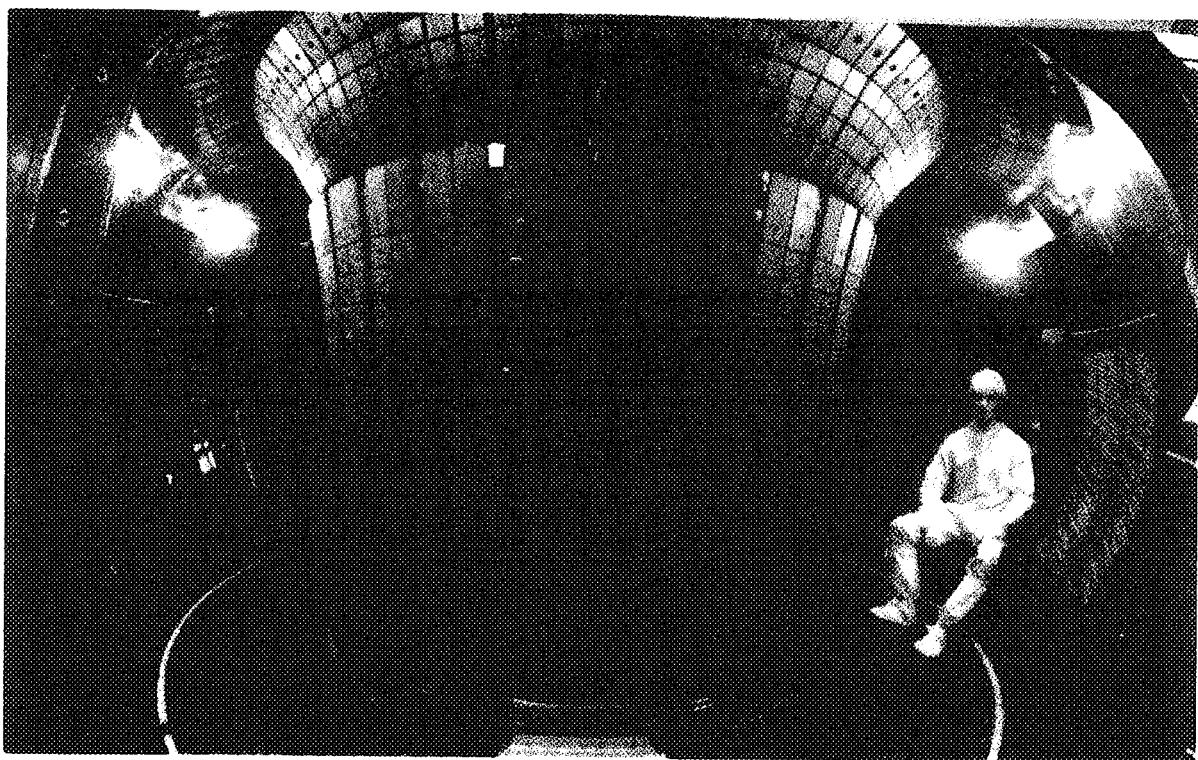
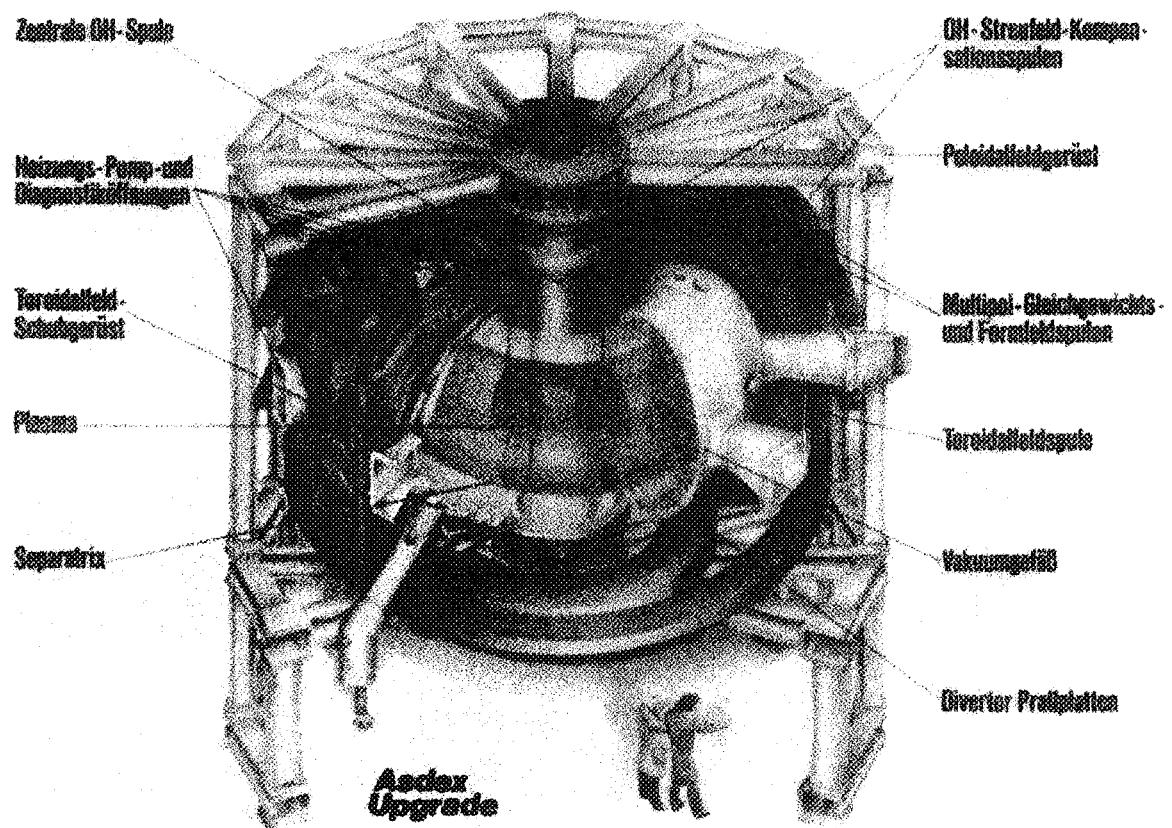
Organized by Ken Hanson, DX-3, 505-667-1402, kmh@lanl.gov

# **Fitting scattered data**

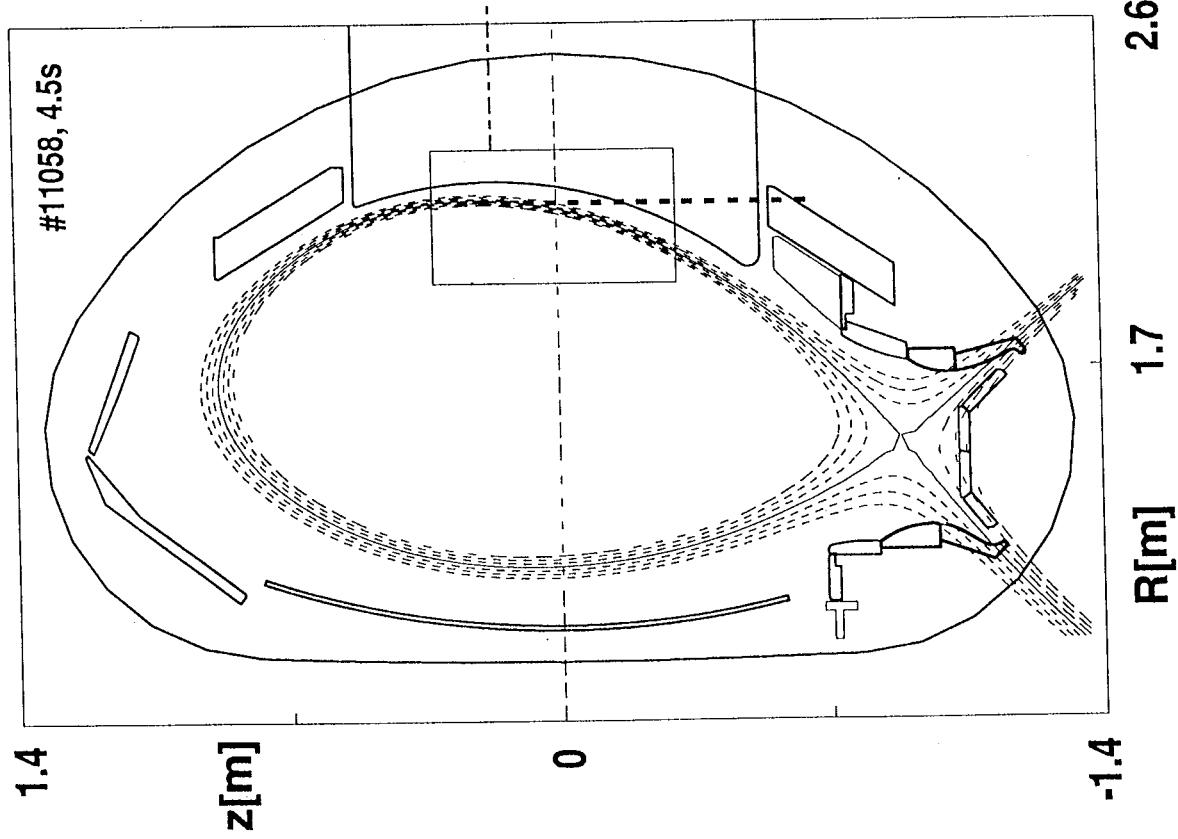
**V. Dose and J. Neuhauser**

**Los Alamos Nat. Lab**

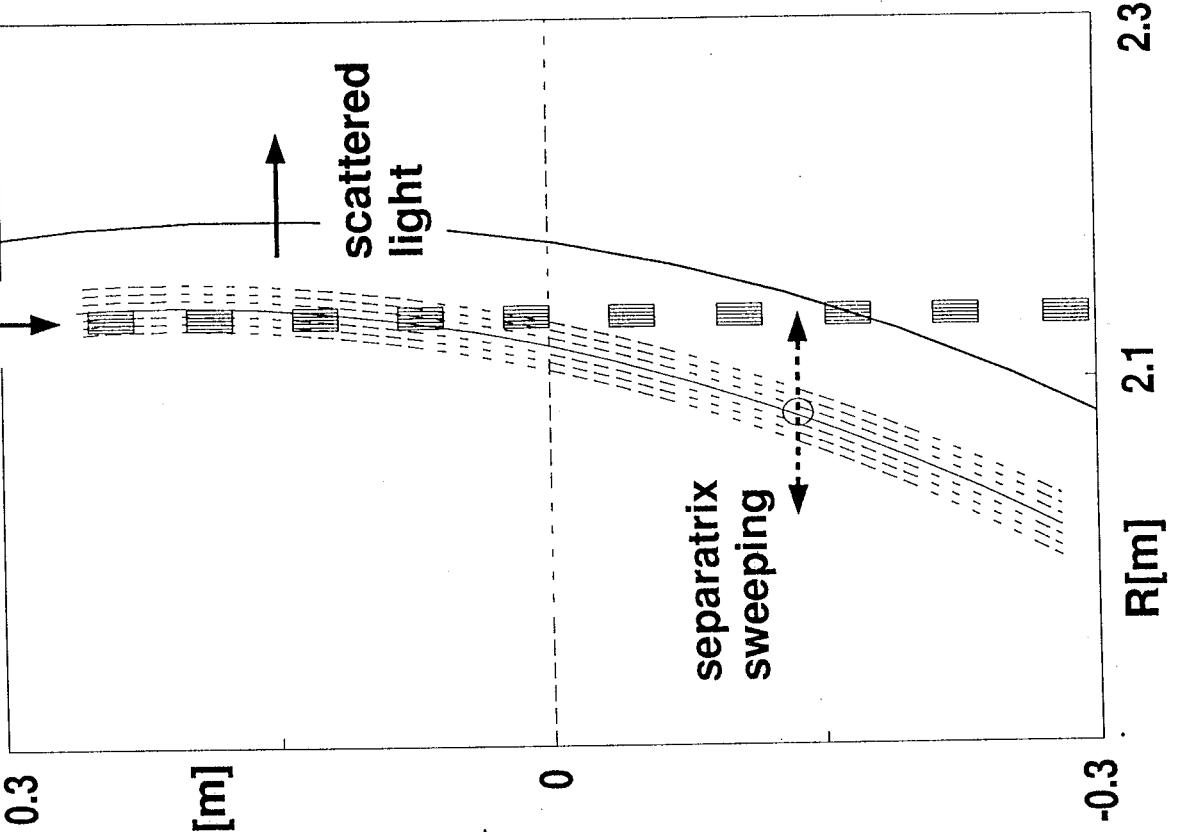
**April 3 – 5, 2000**

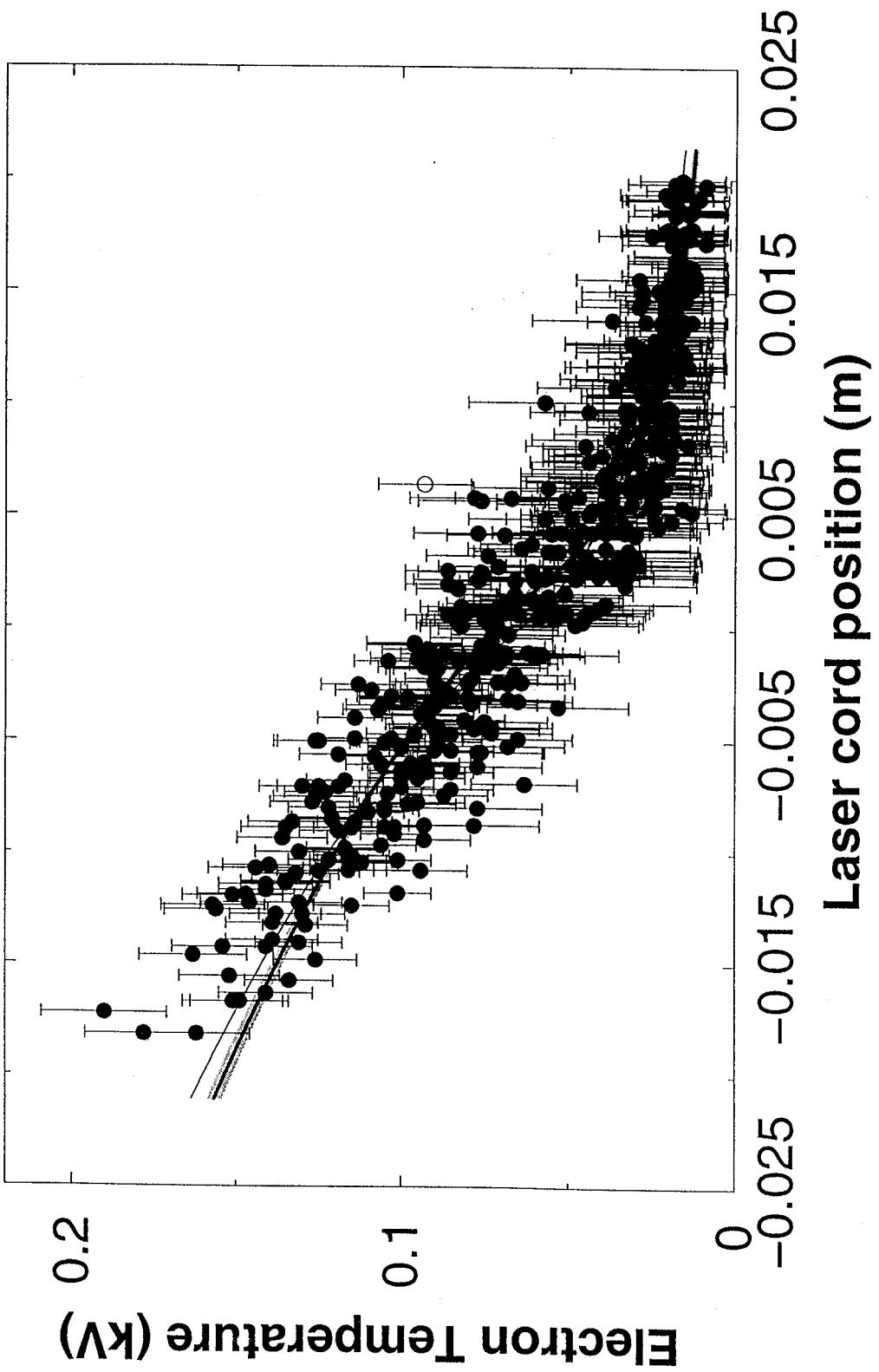


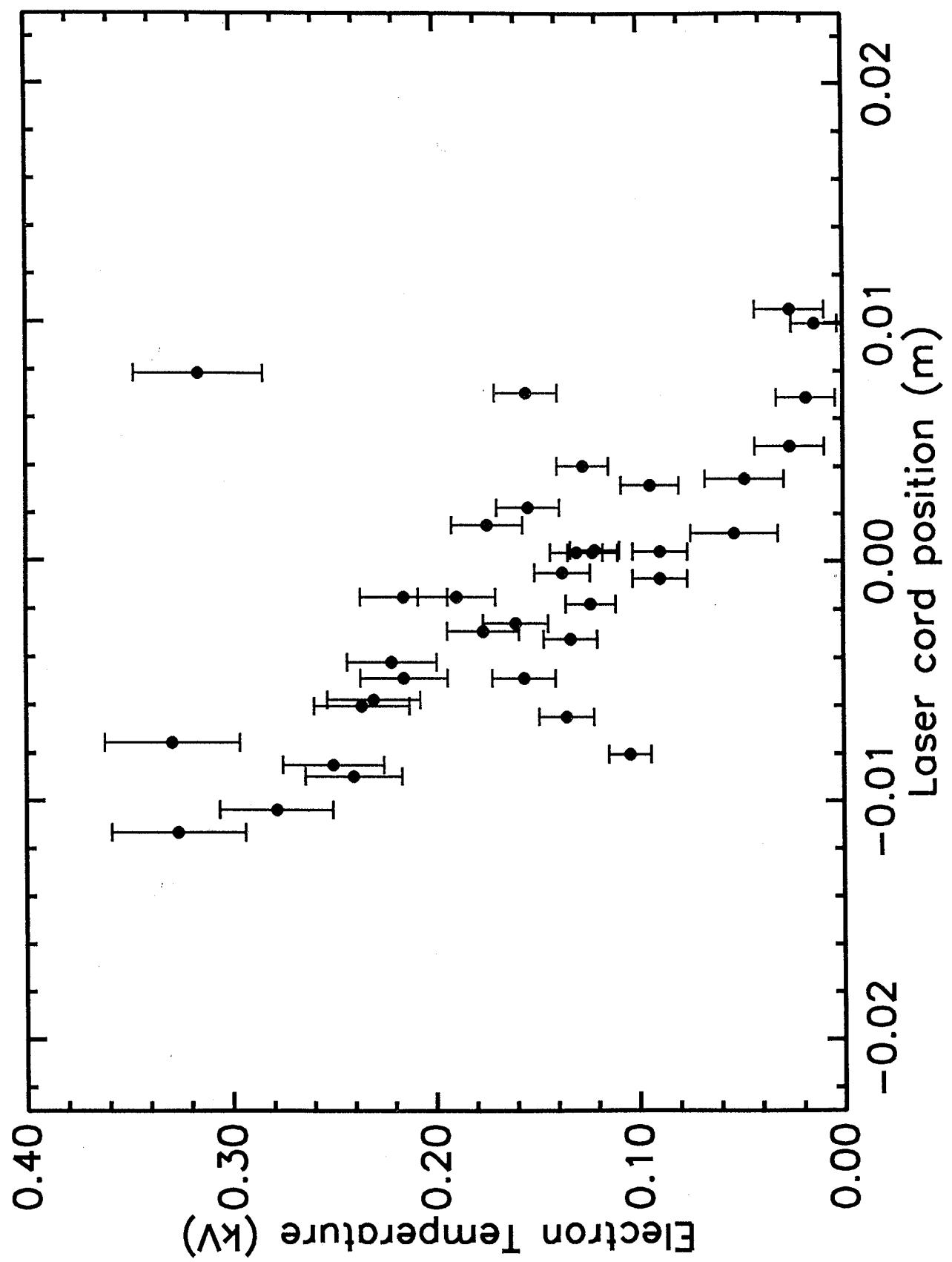
## ASDEX Upgrade



## six parallel laser beams







## model functions

$$\left. \begin{array}{ll} T(x) = \frac{c}{\lambda^{2/7}} \left(1 + \frac{x-x_0}{\lambda}\right)^{-\frac{4}{5}} & x \geq x_0 \\ T(x) = \frac{c}{\lambda^{2/7}} \left(1 - \frac{4}{5} \frac{x-x_0}{\lambda}\right) & x \leq x_0 \end{array} \right\} \begin{array}{l} \text{const} \\ \text{head cond.} \end{array}$$

$$\left. \begin{array}{ll} T(x) = \frac{c}{\lambda^{2/7}} \left(1 + \frac{x-x_0}{\lambda}\right)^{-\frac{4}{3}} & x \geq x_0 \\ T(x) = \frac{c}{\lambda^{2/7}} \left(1 - \frac{8}{3} \frac{x-x_0}{\lambda}\right)^{\frac{1}{2}} & x \leq x_0 \end{array} \right\} \begin{array}{l} \text{head cond.} \\ \sim T \end{array}$$

**likelihood: “regular data”**

$$d_i = T(x_i, c, x_0, \lambda) + \varepsilon_i$$

assume  $\langle \varepsilon_i \rangle = 0$  and  $\langle \varepsilon_i^2 \rangle = \rho_i^2$

$$p(d_i|x_i, c, x_0, \lambda, \rho_i) = \frac{1}{\rho_i \sqrt{2\pi}} \exp \left\{ -\frac{(d_i - T_i)^2}{2\rho_i^2} \right\}$$

$\rho_i$  = true error, unknown, estimate  $\sigma_i$ :

$$p(\rho_i|\sigma_i) = \delta(\rho_i - \sigma_i)$$

## likelihood: “irregular data”

$$d_i = T(x_i, c, x_0, \lambda) + \varepsilon_i \pm s_i$$

$$p^\pm(d_i|x_i, c, x_0, \lambda, \sigma_i, s_i, I) = \frac{1}{\sigma\sqrt{2\pi}} \exp \left\{ -\frac{(d_i - T_i \pm s_i)^2}{2\sigma_i^2} \right\}$$

marginalize  $s_i$

$$\begin{aligned} p^\pm(d_i|x_i, c, x_0, \lambda, \sigma_i, I) &= \int p^\pm(d_i, s_i|x_i, c, x_0, \lambda, \sigma_i, s_i, I) \, ds_i \\ &= \int p^\pm(s_i|I) \cdot p^\pm(d_i|x_i, c, x_0, \lambda, \sigma_i, s_i, I) \, ds_i \end{aligned}$$

assume  $\langle s_i \rangle = \xi$  for all  $i$

$$p^\pm(s_i|\xi, I) = \frac{1}{Z} \exp \left\{ \mp \frac{s_i}{\xi} \right\}$$

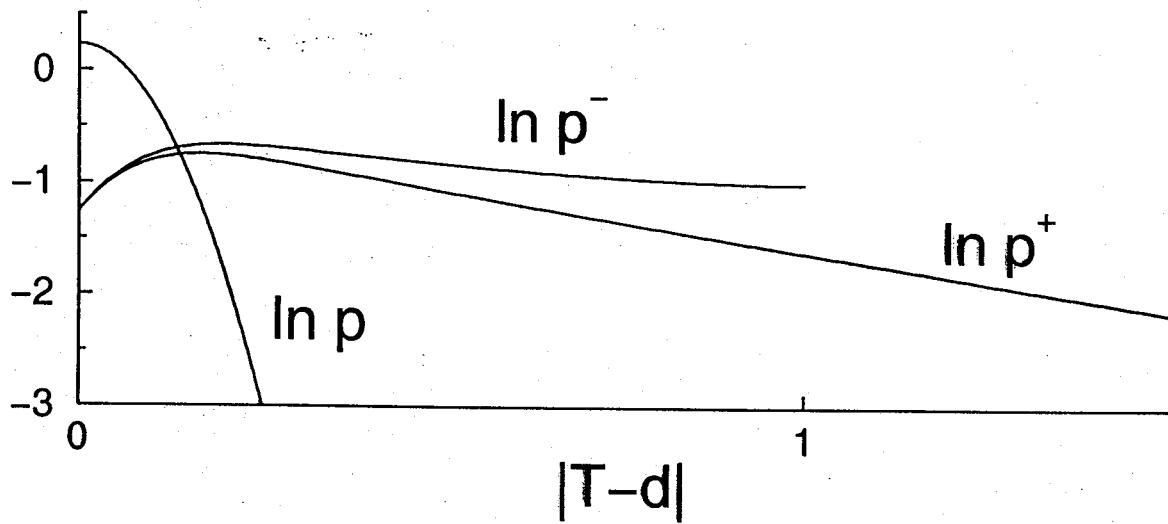
+ sign for  $d > T$  in  $0 \leq d - T < \infty$

- sign for  $d < T$  in  $0 \leq T - d \leq T$

$$Z_i = \xi \left[ 2 - \exp \left( -\frac{T_i}{\xi} \right) \right]$$

$$p^+(d_i|x_i, \sigma_i, T_i, \xi, I) = \exp \left\{ -\frac{d_i - T_i}{\xi} + \frac{\sigma_i^2}{2\xi^2} \right\} \\ \cdot \left\{ \operatorname{erf} \left[ (d_i + \frac{\sigma_i^2}{\xi}) / \sigma_i \sqrt{2} \right] - \operatorname{erf} \left[ (d_i - T_i + \frac{\sigma_i^2}{\xi}) / \sigma_i \sqrt{2} \right] \right\} / Z$$

$$p^-(d_i|x_i, \sigma_i, T_i, \xi, I) = \exp \left\{ \frac{d_i - T_i}{\xi} + \frac{\sigma_i^2}{2\xi^2} \right\} \\ \cdot \left\{ 1 + \operatorname{erf} \left[ (d_i - T_i + \frac{\sigma_i^2}{\xi}) / \sigma_i \sqrt{2} \right] \right\} / Z$$



Mixture model

$$p(\vec{d}|\vec{x}, \vec{\sigma}, \vec{T}, \xi, I) = \prod_i (\beta A_i + (1-\beta)B_i)$$

## Posterior expectations

$$p(c, x_0, \lambda, \beta | \vec{x}, \vec{d}, \vec{\sigma}, I) = \frac{p(c, x_0, \lambda, \beta | I)}{p(\vec{d} | \vec{x}, \vec{\sigma}, I)} p(\vec{d} | \vec{x}, \vec{\sigma}, c, x_0, \lambda, \beta, I)$$

$$p(c, x_0, \lambda, \beta | I) = p(c|I) \cdot p(x_0|I) \cdot p(\lambda|I) \cdot p(\beta|I)$$

$$p(\beta|I) = 1, \quad 0 \leq \beta \leq 1, \quad P(\lambda) = \text{flat, improper}$$

$$p(c|c_0, \Delta c, I) = \exp \left\{ -\frac{1}{2} \left[ \frac{c - c_0}{\Delta c} \right]^2 \right\} / \Delta c \sqrt{2\pi}$$

$$p(x_0|\Delta x_0, I) = \exp \left\{ -\frac{1}{2} \left[ \frac{x_0}{\Delta x_0} \right]^2 \right\} / \Delta x_0 \sqrt{2\pi}$$

## Parameters

$$\langle \mu_k^n \rangle = \int dc \int dx_0 \int d\lambda \int d\beta \mu_k^n p(c, x_0, \lambda, \beta | \vec{x}, \vec{d}, \vec{\sigma}, I)$$

## Function

$$\langle T^n(x) \rangle = \int [T(x, x_0, c, \lambda)]^n \cdot p(c, x_0, \lambda, \beta | \vec{x}, \vec{d}, \vec{\sigma}, I) dc dx_0 d\lambda d\beta$$

